# **Boundary Value Problem**

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### **Overview**

- BVP in 1D
- Boundary Conditions (BC)
- Solution Methods
  - Finite-Diference Method (FDM)
  - Shooting Method
- Steps in FDM
- FDM for General Linear Equation in 1D
- FDM for General Nonlinear Equation in 1D

### **BVP in 1D**

Problem: for a given function f(x,y,z) find y(x) such that

$$\frac{d^2y}{dx^2} = y'' = f(x, y, y') \qquad a < x < b \quad \text{numbers } \alpha \text{ and } \beta$$

$$\text{are given.}$$

$$y(a) = \alpha \quad y(b) = \beta$$

The problem is charecterized by:

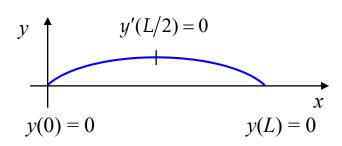
- > Second order differential equation;
- $\triangleright$  Two boundary values, y(a) and y(b), instead of initial values, y(a), y'(a);
- Independent variable x is more naturally interpreted as a coordinate in space rather than in time;
- ➤ Big difference in mathematical and numerical treatment;
- In applications, function f is typically linear, f = a(x)y' + b(x)y + c(x)

# **Example Equation**

$$y'' = f(x) \qquad 0 < x < L$$
$$y(0) = 0 \qquad y(L) = 0$$

#### Boundary condition (BC)

• Dirichlet, 1<sup>st</sup> type:  $y(0) = \alpha$ 



- Neumann (temperature insulation), 2<sup>nd</sup> type:
  - $y'(0) = 0 \ [y'(0) = \alpha]$
- Robin (cooling),  $3^{rd}$  type:  $y'(0) = \alpha_1 y(0) + \alpha_2$

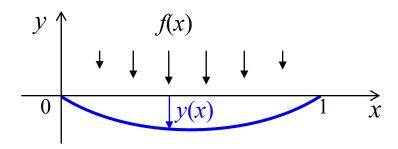


Newton's law of cooling  $T'(L) = -k[T(L) - T_0]$ 

# **Physical Interpretation**

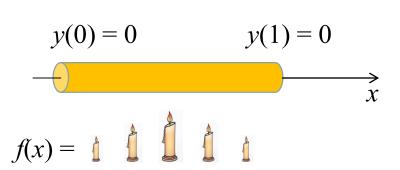
#### Elastic string

- y = displacement
- f =force distribution



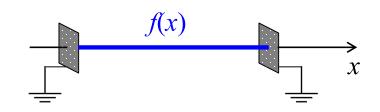
#### Heat conduction

- y = temperature
- f = heat source



#### Electrostatic

- y = potential
- f =charge density



# Warning

$$y'' = a(x)y' + b(x)y + c(x) \qquad a < x < b$$
$$y(a) = \alpha \qquad y(b) = \beta$$

Solution existence and uniqueness for boundary value problems are more involved questions than for initial value problem.

**Example:** y'' = -y with y(0) = 0 and  $y(2\pi) = 1$  has no solution.

### **Linear BVP**

$$\begin{cases} y'' = a(x)y' + b(x)y + c(x) & a < x < b \\ y(a) = \alpha & y(b) = \beta \end{cases}$$

For a linear boundary value problem, there exists a unique solution when a(x), b(x) and c(x) are continuous functions and  $b(x) \ge 0$ .

$$f(x, y, z) = a(x)z + b(x)y + c(x)$$

### **Non-Linear BVP**

Given 
$$f(x, y, z)$$
 
$$\begin{cases} y'' = f(x, y, y') \\ y(a) = \alpha \quad y(b) = \beta \end{cases}$$

For this boundary value problem, assume that f(x, y, z) satisfies

- the partial derivatives f<sub>x</sub>, f<sub>y</sub>, f<sub>z</sub> are continuous
   f<sub>y</sub>(t, y, z) > 0 and
   |f<sub>z</sub>(t, y, z)| ≤ M for some M when x ∈ [a,b] and all y and z.

Then the BVP problem has a unique solution y(x).

### **Solution Methods for BVPs**

$$y'' = a(x)y' + b(x)y + c(x) \qquad a < x < b$$
$$y(a) = \alpha \qquad y(b) = \beta$$

We know how to solve if y'(a) = z.

#### Several methods:

- (1) Shooting method, we find y'(a) = z iteratively;
- (2) Finite difference method, FDM, (matrix method);
- (3) Collocation method;
- (4) Finite Element, Galerkin Method.

### **FD Method**

#### 1. Discretization

$$x \longrightarrow [x_0, x_1, \dots, x_{n+1}]^T \equiv \mathbf{x}$$

$$y(x) \longrightarrow [y_0, y_1, \dots, y_{n+1}]^T \equiv \mathbf{y}$$

2. Approximation

$$\frac{dy(x_i)}{dx} \longrightarrow \frac{y_{i+1} - y_i}{x_{i+1} - x_i}$$

3. Solving FD equations

$$\mathbf{A}\mathbf{y} = \mathbf{f}$$

4. Answering the question

$$|y(x_i) - y_i| \le Ch^m$$
  $h = \max_i (x_{i+1} - x_i)$ 

# Step 1 in FDM

$$y''(x) = f(x) \quad a < x < b$$

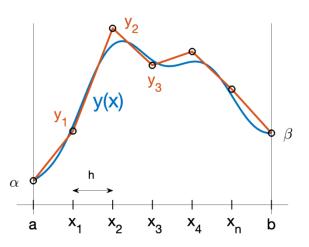
$$y''(x) = f(x)$$
  $a < x < b$   
 $y(a) = \alpha$   $y(b) = \beta$ 

### (1) Discretization

$$h = \frac{b-a}{n+1}$$
  $x_j = a+j \cdot h$   $j = 0,1,2,...,n+1$ 

Approximate exact solution in  $x_i$  with  $y_i$ 

$$y_j \approx y(x_j)$$

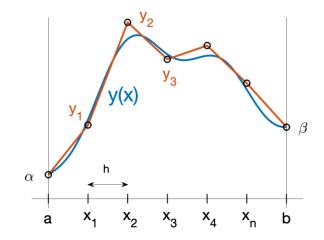


# Step 2 in FDM

$$y''(x) = f(x) \quad a < x < b$$

$$y''(x) = f(x)$$
  $a < x < b$   
 $y(a) = \alpha$   $y(b) = \beta$ 

#### (2) Numerical differentiation



$$y''(x_j) = \frac{y(x_{j-1}) - 2y(x_j) + y(x_{j+1})}{h^2} + O(h^2) \quad j = 1, 2, ..., n$$

It gives

$$\frac{y(x_{j-1}) - 2y(x_j) + y(x_{j+1})}{h^2} = f(x_j) + O(h^2) \quad j = 1, 2, ..., n$$

# Step 3 in FDM

We have for the moment

$$\frac{y(x_{j-1}) - 2y(x_j) + y(x_{j+1})}{h^2} = f(x_j) + O(h^2) \quad j = 1, 2, ..., n$$

(3) Set FD equations neglecting  $O(h^2)$  and  $y(x_i) \rightarrow y_i$ 

$$\frac{y_{j-1} - 2y_j + y_{j+1}}{h^2} = f(x_j) \quad j = 1, 2, \dots, n$$

It defines numbers  $y_j$  j = 1,2, ..., n

# Step 4 in FDM

#### Given *n* equations

$$\frac{y_{j-1} - 2y_j + y_{j+1}}{h^2} = f(x_j) \quad j = 1, 2, \dots, n$$

#### (4) Using boundary conditions

$$j = 1 \to \frac{y_0 - 2y_1 + y_2}{h^2} = f(x_1) \qquad \frac{-2y_1 + y_2}{h^2} = f(x_1) - \frac{\alpha}{h^2}$$

$$j = n \rightarrow \frac{y_{n-1} - 2y_n + y_{n+1}}{h^2} = f(x_n) \qquad \frac{y_{n-1} - 2y_n}{h^2} = f(x_n) - \frac{\beta}{h^2}$$

#### There are n unknowns, $y_i$ , and n equations

## Step 5 in FDM

We have two systems of equations

$$\frac{y(x_{j-1}) - 2y(x_j) + y(x_{j+1})}{h^2} = f(x_j) + r_j \qquad r_j = O(h^2)$$

$$\frac{y_{j-1} - 2y_j + y_{j+1}}{h^2} = f(x_j) \qquad j = 1, 2, ..., n$$

(5) Subtracting and defining error  $e_j = y(x_j) - y_j$ 

$$\frac{e_{j-1} - 2e_j + e_{j+1}}{h^2} = r_j = O(h^2) \qquad j = 1, 2, ..., n$$

It may be proven,  $e_i = O(h^2)$ 

## Step 6 in FDM

$$\frac{-2y_1 + y_2}{h^2} = f(x_1) - \frac{\alpha}{h^2}$$

$$\frac{y_{j-1} - 2y_j + y_{j+1}}{h^2} = f(x_j) \qquad j = 1, 2, ..., n$$

$$\frac{y_{n-1} - 2y_n}{h^2} = f(x_n) - \frac{\beta}{h^2}$$

(6) Rewriting in matrix form Ay = b  $y, b \in \mathbb{R}^n$   $A \in \mathbb{R}^{n \times n}$ 

$$\mathbf{A}\mathbf{y} = \mathbf{b} \quad \mathbf{y}, \mathbf{b} \in \mathbb{R}^n$$

$$\mathbf{A} \in \mathbf{R}^{n \times n}$$

$$\mathbf{A} = \frac{1}{h^{2}} \begin{bmatrix} -2 & 1 & & & \\ 1 & -2 & 1 & & \\ & \ddots & \ddots & \ddots & \\ & & 1 & -2 & 1 \\ & & & 1 & -2 \end{bmatrix} \qquad \mathbf{y} = \begin{bmatrix} y_{1} \\ y_{2} \\ \vdots \\ y_{n-1} \\ y_{n} \end{bmatrix} \qquad \mathbf{b} = \begin{bmatrix} f(x_{1}) - \alpha/h^{2} \\ f(x_{2}) \\ \vdots \\ f(x_{n}) - \beta/h^{2} \end{bmatrix}$$

$$\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_{n-1} \\ y_n \end{bmatrix}$$

$$\mathbf{b} = \begin{bmatrix} f(x_1) - \alpha/h^2 \\ f(x_2) \\ \vdots \\ f(x_{n-1}) \\ f(x_n) - \beta/h^2 \end{bmatrix}$$

# **Diagonally Dominant**

Matrix  $\mathbf{A} = [a_{ii}]$  is said diagonally dominant if

$$|a_{ii}| \ge \sum_{j \ne i} |a_{ij}|$$
  $\forall i$  and at least one inequality is strict

$$\mathbf{A} = \frac{1}{h^2} \begin{bmatrix} -2 & 1 \\ 1 & -2 & 1 \\ & \ddots & \ddots \\ & & 1 & -2 & 1 \\ & & & 1 & -2 & 1 \\ & & & & 1 & -2 \end{bmatrix}$$
 Diagonally dominant matrices  
• Non-singular  
• Numerically stable under LU  
• Need no pivoting/scaling  
• Often arise in FDM

## **More Accurate Approximation**

$$y_i'' = \frac{-y_{i-2} + 16y_{i-1} - 30y_i + 16y_{i+1} - y_{i+2}}{12h^2} + \frac{y^{(5)}(\xi)}{90}h^4$$

### Matrix Form

Finite difference method leads to linear equations Ay = b

$$\mathbf{A} = \frac{1}{h^{2}} \begin{bmatrix} -2 & 1 & & & \\ 1 & -2 & 1 & & \\ & \ddots & \ddots & \ddots & \\ & & 1 & -2 & 1 \\ & & & 1 & -2 \end{bmatrix} \qquad \mathbf{y} = \begin{bmatrix} y_{1} \\ y_{2} \\ y_{3} \\ \vdots \\ y_{n} \end{bmatrix} \qquad \mathbf{b} = \begin{bmatrix} f(x_{1}) - \alpha/h^{2} \\ f(x_{2}) \\ f(x_{3}) \\ \vdots \\ f(x_{n}) - \beta/h^{2} \end{bmatrix}$$

$$\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_n \end{bmatrix}$$

$$\mathbf{b} = \begin{bmatrix} f(x_1) - \alpha/h^2 \\ f(x_2) \\ f(x_3) \\ \vdots \\ f(x_n) - \beta/h^2 \end{bmatrix}$$

- Matrices are sparse; use sparse format in Matlab
- When y'' is properly approximated by FD, error is  $O(h^2)$

$$\max_{1 \le j \le n} \left| y(x_j) - y_j \right| \le Ch^2$$

## **Example Code**

```
N = 80;
                                 % Discretization parameters
h = (b-a)/(N+1);
                                 % Discretization step
                                 % Vector of inner nodes/points
x = a:h:b;
ettor = ones(N, 1);
                                 % Construct matrix A
      = -2*diag(ettor)+...
           diag(ettor(1:end-1),1)+...
           diag(ettor(1:end-1),-1);
      = sparse(A)/h^2;
Α
                                % f = RHS function
bvec = f(x);
bvec(1) = bvec(1)-alpha/h^2; % Correction for left BC
bvec(end) = bvec(end)-beta/h^2; % Correction for right BC
          = A \setminus bvec;
                                 % y = Solution in inner nodes
У
```

### **More Efficient Code in 1D**

```
N = 80;
                                % Discretization parameters
                                % Discretization step
h = (b-a)/(N+1);
x = a:h:b;
                                % Vector of inner nodes/points
A = ones(N-1,1)/h^2;
                                % Super/Sub-diagonal
B = -2*ones(N, 1)/h^2;
                                % Diagonal
    = f(x);
bvec
                                % RHS
bvec(1) = bvec(1)-alpha/h^2; % Left BC correction
bvec(end) = bvec(end) -beta/h^2; % Right BC correction
y = tridisolve(A, B, A, bvec); % y = solution in inne nodes
```

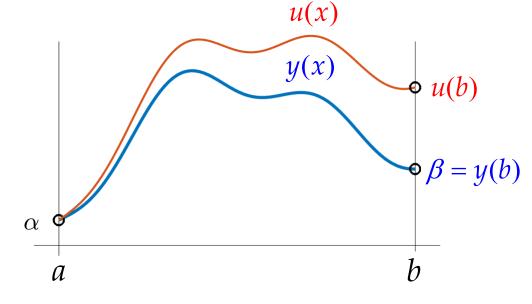
```
function x = tridisolve(a,b,c,d)
% TRIDISOLVE Solve tridiagonal system of equations.
% x = TRIDISOLVE(a,b,c,d) solves the system of linear equations
                   b(1) *x(1) + c(1) *x(2) = d(1),
% a(j-1)*x(j-1) + b(j)*x(j) + c(j)*x(j+1) = d(j), j = 2:n-1,
% a(n-1)*x(n-1) + b(n)*x(n)
                                             = d(n).
% The algorithm does not use pivoting, so the results might
% be inaccurate if abs(b) is much smaller than abs(a)+abs(c).
% More robust, but slower, alternatives with pivoting are:
% x = T \setminus d \text{ where } T = diag(a, -1) + diag(b, 0) + diag(c, 1)
% x = S d \text{ where } S = \text{spdiags}([[a; 0] b [0; c]], [-1 0 1], n, n)
% Copyright 2014 Cleve Moler (The MathWorks, Inc)
x = d;
n = length(x);
for j = 1:n-1
    mu = a(i)/b(i);
    b(j+1) = b(j+1) - mu*c(j);
    x(j+1) = x(j+1) - mu*x(j);
end
x(n) = x(n)/b(n);
for j = n-1:-1:1
    x(i) = (x(i)-c(i)*x(i+1))/b(i);
end
```

# **Shooting Method**

Again, we start with the BVP

$$y''(x) = f(x)$$
  $a < x < b$   
 $y(a) = \alpha$   $y(b) = \beta$ 

Instead, we formulate an IVP with a parameter z



- u''(x) = f(x) a < x < b $u(a) = \alpha u'(a) = z$
- If the solution u(x) satisfies  $u(b) = \beta$  then  $u \equiv y$  (uniqueness theorem);
- The problem is to find z.

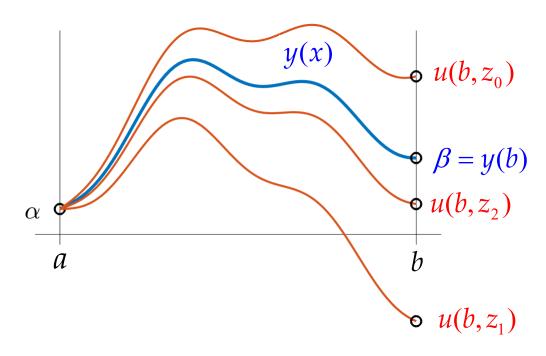
# **Shooting Method Essence**

#### **Initial Value Problem**

$$u''(x) = f(x) \qquad a < x < b$$

$$u(a) = \alpha$$

$$u'(a) = z$$



# **Shooting Method Algorithm**

- Let u(x,z) be a solution to IVP;
- We are trying to find z such that  $u(b,z) = \beta$ ;
- Introduce function,  $G(z) \equiv u(b,z) \beta$  so z becomes root for G(z);
- We evaluate G(z) approximately by solving (IVP) numerically;
- Use the secant method to solve G(z) = 0.

$$z_{n+1} = z_n - \frac{z_n - z_{n-1}}{\tilde{G}(z_n) - \tilde{G}(z_{n-1})} \tilde{G}(z_n) \quad [\text{Numerical evaluation of } G(z)]$$

### Remarks

- ➤ Combination of 2 solvers: ODE and (nonlinear) solver;
- ➤ Works equally good for both linear and nonlinear eqs;
- ➤ Works only in 1D and typically for 2nd order eqs;
- > Requires initial guess as contrast to FDM;
- > Typically, FDM is more effective and works as well in 2D and 3D.

### **FD Method**

General linear equation

$$y'' = a(x)y' + b(x)y + c(x) \qquad 0 < x < 1$$
$$y(0) = \alpha \qquad y(1) = \beta$$

• Neumann/Robin boundary condition

$$y'(0) = \alpha$$
  $y'(0) = \alpha_1 y(0) + \alpha_2$ 

• Nonlinear equations

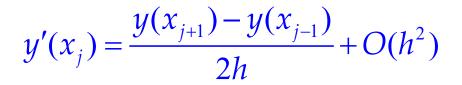
$$y'' = F(x, y, y') \qquad 0 < x < 1$$
$$y(0) = \alpha \qquad y(1) = \beta$$

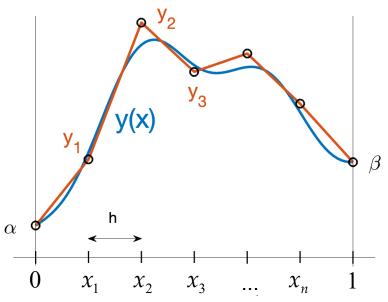
## FDM for General Lin Eq

$$y'' = a(x)y' + b(x)y + c(x) \qquad 0 < x < 1$$
$$y(0) = \alpha \qquad y(1) = \beta$$

- (1) Discretize as before
- (2) Approximate y'' and y'

$$y'(x_j) = \frac{y(x_j) - y(x_{j-1})}{h} + O(h)$$





$$h = 1/(n+1)$$
  
 $x_j = jh$   $j = 0, 1, ..., N+1$ 

# **Setting FD Equations**

$$y'' - a(x)y' - b(x)y = c(x)$$

(3) Collecting FDs in one equation

$$\frac{y(x_{j-1}) - 2y(x_j) + y(x_{j+1})}{h^2} - a(x_j) \frac{y(x_{j+1}) - y(x_{j-1})}{2h} - b(x_j)y(x_j) = c(x_j) + O(h^2)$$

(4) Neglecting  $O(h^2)$  and  $y(x_j) \rightarrow y_j$ 

$$\frac{y_{j-1} - 2y_j + y_{j+1}}{h^2} - a(x_j) \frac{y_{j+1} - y_{j-1}}{2h} - b(x_j) y_j = c(x_j)$$

## **Collecting Similar Terms**

$$\frac{y_{j-1} - 2y_j + y_{j+1}}{h^2} - a(x_j) \frac{y_{j+1} - y_{j-1}}{2h} - b(x_j) y_j = c(x_j)$$

$$\left[\frac{1}{h^{2}} + \frac{a(x_{j})}{2h}\right] y_{j-1} + \left[-\frac{2}{h^{2}} - b(x_{j})\right] y_{j} + \left[\frac{1}{h^{2}} - \frac{a(x_{j})}{2h}\right] y_{j+1} = c(x_{j})$$

Rewriting in compact form

$$p_{j}y_{j-1} + q_{j}y_{j} + r_{j}y_{j+1} = c(x_{j})$$

$$j = 1, 2, ..., n$$

$$p_{j} = \frac{1}{h^{2}} + \frac{a(x_{j})}{2h}$$

$$q_{j} = -\frac{2}{h^{2}} - b(x_{j})$$

$$r_{j} = \frac{1}{h^{2}} - \frac{a(x_{j})}{2h}$$

# **Using BCs**

#### Given

$$p_{j}y_{j-1} + q_{j}y_{j} + r_{j}y_{j+1} = c(x_{j})$$

$$j = 1, 2, ..., n$$

$$p_{j} = \frac{1}{h^{2}} + \frac{a(x_{j})}{2h}$$

$$q_{j} = -\frac{2}{h^{2}} - b(x_{j})$$

$$r_{j} = \frac{1}{h^{2}} - \frac{a(x_{j})}{2h}$$

• When  $j = 1, y_0 = \alpha$ 

$$p_1 y_0 + q_1 y_1 + r_1 y_2 = c(x_1) \rightarrow q_1 y_1 + r_1 y_2 = c(x_1) - p_1 \alpha$$

• When  $j = n, y_{n+1} = \beta$ 

$$p_n y_{n-1} + q_n y_n + r_n y_{n+1} = c(x_n) \to p_n y_{n-1} + q_n y_n = c(x_n) - r_n \beta$$

We have *n* unknowns and *n* equations!

### **Matrix Form**

$$q_{1}y_{1} + r_{1}y_{2} = c(x_{1}) - p_{1}\alpha \qquad j = 1$$
We arrive at 
$$p_{j}y_{j-1} + q_{j}y_{j} + r_{j}y_{j+1} = c(x_{j}) \qquad j = 2, ..., n-1$$

$$p_{n}y_{n-1} + q_{n}y_{n} = c(x_{j}) - r_{n}\beta \qquad j = n$$

#### Rewriting in matrix form

$$\mathbf{A} = \begin{bmatrix} q_1 & r_1 \\ p_2 & q_2 & r_2 \\ & \ddots & \ddots & \ddots \\ & & p_{n-1} & q_{n-1} & r_{n-1} \\ & & & p_n & q_n \end{bmatrix} \quad \mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_{n-1} \\ y_n \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} c(x_1) - p_1 \alpha \\ c(x_2) \\ \vdots \\ c(x_{n-1}) \\ c(x_n) - r_n \beta \end{bmatrix}$$

$$\left| y(x_j) - y_j \right| \le C \cdot h^2$$

### **Neumann BC**

Instead of Dirichlet  $y(0) = \alpha$ We use Neumann  $y'(0) = \alpha$ 

Previous steps are same. Only distinction: how to use BC ( $x_0 = 0$ )

$$y'(0) = y'(x_0) = \frac{y(x_1) - y(x_0)}{h} + O(h)$$

• Neglect O(h) and replace  $y(x_j) \rightarrow y_j$ 

$$\frac{y_1 - y_0}{h} = y'(0) = \alpha \to y_0 = y_1 - h\alpha$$

$$p_1y_0 + q_1y_1 + r_1y_2 = c(x_1) \longrightarrow (p_1 + q_1)y_1 + r_1y_2 = c(x_1) + p_1h\alpha$$

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## **Matrix Form again**

$$(p_1 + q_1)y_1 + r_1y_2 = c(x_1) + p_1h\alpha \qquad j = 1$$
Now we have 
$$p_j y_{j-1} + q_j y_j + r_j y_{j+1} = c(x_j) \qquad j = 2, ..., n-1$$

$$p_n y_{n-1} + q_n y_n = c(x_j) - r_n \beta \qquad j = n$$

#### Rewriting in matrix form

$$\mathbf{A} = \begin{bmatrix} p_{1} + q_{1} & r_{1} & & & & \\ p_{2} & q_{2} & r_{2} & & & \\ & \ddots & \ddots & \ddots & & \\ & & p_{n-1} & q_{n-1} & r_{n-1} \\ & & & p_{n} & q_{n} \end{bmatrix} \quad \mathbf{y} = \begin{bmatrix} y_{1} \\ y_{2} \\ \vdots \\ y_{n-1} \\ y_{n} \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} c(x_{1}) + p_{1}h\alpha \\ c(x_{2}) \\ \vdots \\ c(x_{n-1}) \\ c(x_{n}) - r_{n}\beta \end{bmatrix}$$

$$|y(x_j) - y_j| \le C \cdot h$$
 Can be improved!

# Restoring $O(h^2)$

$$y'' = a(x)y' + b(x)y + c(x) y(h) = y(0) + hy'(0) + \frac{h^2}{2!}y''(0) + \frac{h^3}{3!}y'''(\xi)$$

$$\frac{y(h) - y(0)}{h} = y'(0) + \frac{h}{2}y''(0) + O(h^2)$$

$$\frac{y(h) - y(0)}{h} = y'(0) + \frac{h}{2}[a(0)y'(0) + b(0)y(0) + c(0)] + O(h^2)$$

$$\frac{y_1 - y_0}{h} = \alpha + \frac{h}{2}[a(0)\alpha + b(0)y_0 + c(0)]$$

$$\frac{y_1 - y_0}{h} - \frac{hb(0)}{2}y_0 = \alpha \left[1 + \frac{ha(0)}{2}\right] + \frac{hc(0)}{2}$$

# FD for Nonlinear Eq

#### Nonlinear problem

$$y''(x) = F(x, y(x))$$
  $a < x < b$   $h = (b-a)/(n+1)$   
 $y(a) = \alpha$   $y(b) = \beta$   $x_j = a + jh$   $j = 0, 1, ..., n+1$ 

#### Similar steps as before

$$\frac{\alpha - 2y_1 + y_2}{h^2} = F(x_1, y_1) \qquad j = 1$$

$$\frac{y_{j-1} - 2y_j + y_{j+1}}{h^2} = F(x_j, y_j) \qquad j = 2, \dots, n-1$$

$$\frac{y_{n-1} - 2y_n + \beta}{h^2} = F(x_n, y_n) \qquad j = n$$

$$\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

#### Nonlinear now!

### **Vector-Valued Form**

#### Given

$$\frac{\alpha - 2y_1 + y_2}{h^2} = F(x_1, y_1) \qquad j = 1$$

$$\frac{y_{j-1} - 2y_j + y_{j+1}}{h^2} = F(x_j, y_j) \qquad j = 2, \dots, n-1$$

$$\frac{y_{n-1} - 2y_n + \beta}{h^2} = F(x_n, y_n) \qquad j = n$$

$$f_{1}(\mathbf{y}) \equiv (\alpha - 2y_{1} + y_{2})/h^{2} - F(x_{1}, y_{1})$$

$$f_{j}(\mathbf{y}) \equiv (y_{j-1} - 2y_{j} + y_{j+1})/h^{2} - F(x_{j}, y_{j})$$

$$f_{n}(\mathbf{y}) \equiv (y_{n-1} - 2y_{n} + \beta)/h^{2} - F(x_{n}, y_{n})$$

$$\mathbf{F}(\mathbf{y}) \equiv \begin{bmatrix} f_{1}(\mathbf{y}) \\ f_{2}(\mathbf{y}) \\ \vdots \\ f_{n}(\mathbf{y}) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\mathbf{F}(\mathbf{y}) \equiv \begin{bmatrix} f_1(\mathbf{y}) \\ f_2(\mathbf{y}) \\ \vdots \\ f_n(\mathbf{y}) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\mathbf{F}(\mathbf{y}) = \mathbf{0}$$

### **Vector-Valued Function**

$$f_{1}(\mathbf{y}) \equiv (\alpha - 2y_{1} + y_{2})/h^{2} - F(x_{1}, y_{1}) \qquad f_{1}(\mathbf{y}) \equiv (-2y_{1} + y_{2})/h^{2} \qquad + \alpha/h^{2} - F(x_{1}, y_{1})$$

$$f_{j}(\mathbf{y}) \equiv (y_{j-1} - 2y_{j} + y_{j+1})/h^{2} - F(x_{j}, y_{j}) \qquad f_{j}(\mathbf{y}) \equiv (y_{j-1} - 2y_{j} + y_{j+1})/h^{2} - F(x_{j}, y_{j})$$

$$f_{n}(\mathbf{y}) \equiv (y_{n-1} - 2y_{n} + \beta)/h^{2} - F(x_{n}, y_{n}) \qquad f_{n}(\mathbf{y}) \equiv (y_{n-1} - 2y_{n})/h^{2} \qquad + \beta/h^{2} - F(x_{n}, y_{n})$$

$$\mathbf{F}(\mathbf{y}) = \frac{1}{h^{2}} \begin{bmatrix} -2 & 1 & & & \\ 1 & -2 & 1 & & \\ & \ddots & \ddots & \ddots & \\ & & 1 & -2 & 1 \\ & & & 1 & -2 \end{bmatrix} \cdot \begin{bmatrix} y_{1} \\ y_{2} \\ \vdots \\ y_{n-1} \\ y_{n} \end{bmatrix} - \begin{bmatrix} F(x_{1}, y_{1}) - \alpha/h^{2} \\ F(x_{2}, y_{2}) \\ \vdots \\ F(x_{n-1}, y_{n-1}) \\ F(x_{n}, y_{n}) - \beta/h^{2} \end{bmatrix}$$

### **Jacobian**

$$\mathbf{J}(\mathbf{y}) = \begin{bmatrix} \partial_1 f_1 & \partial_2 f_1 & \cdots & \partial_n f_1 \\ \partial_1 f_2 & \partial_2 f_2 & \cdots & \partial_n f_2 \\ \vdots & \vdots & \ddots & \vdots \\ \partial_1 f_n & \partial_2 f_n & \cdots & \partial_n f_n \end{bmatrix}$$

$$\mathbf{J}(\mathbf{y}) = \begin{bmatrix} \partial_1 f_1 & \partial_2 f_1 & & & 0 \\ \partial_1 f_2 & \partial_2 f_2 & \partial_3 f_2 & & \\ & \ddots & \ddots & \ddots & \\ & & \partial_{n-2} f_{n-1} & \partial_{n-1} f_{n-1} & \partial_n f_{n-1} \\ 0 & & & \partial_{n-1} f_n & \partial_n f_n \end{bmatrix}$$

# **Tridiagonal Jacobian**

$$\mathbf{J}(\mathbf{y}) = \mathbf{A} - \begin{bmatrix} F_{y_1}(x_1, y_1) & 0 \\ F_{y_2}(x_2, y_2) & \\ 0 & F_{y_n}(x_n, y_n) \end{bmatrix}$$

### **Newton's Method**

$$\mathbf{y}^{(k+1)} = \mathbf{y}^{(k)} - \mathbf{J} \left( \mathbf{y}^{(k)} \right)^{-1} \mathbf{F} \left( \mathbf{y}^{(k)} \right)$$

#### Tridiagonal Jacobian

- sparse format
- tridisolve

### **MATLAB Tridisolve**

```
function x = tridisolve(a,b,c,f)
x = f;
n = length(x);
for j = 1:n-1
   mu = a(j)/b(j);
   b(j+1) = b(j+1) - mu*c(j);
   x(j+1) = x(j+1) - mu*x(j);
end
x(n) = x(n)/b(n);
for j = n-1:-1:1
   x(j) = (x(j)-c(j)*x(j+1))/b(j);
end
```

Cost = O(n)

# **Important**

- BVP in 1D
- Boundary Conditions (BC)
- Solution Methods
  - Finite-Diference Method (FDM)
  - Shooting Method
- Steps in FDM
- FDM for General Linear Equation in 1D
- FDM for General Nonlinear Equation in 1D