

Lesson 3

Boundary Value Problems for ODEs

3.1 Well-posedness

3.2 Shooting Method

3.3 Finite Difference Method



Boundary Value Problems: Finite Difference Method



In the shooting method, we start by approximately satisfying the ODE (using an IVP solver) and iterate until the boundary conditions are satisfied.

Alternatively, we can start by satisfying the boundary conditions and iterate until the ODE is satisfied approximately.

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with boundary conditions

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We replace the derivatives appearing in the ODE by finite difference approximations

$$u'(t_i) \approx \frac{u_{i+1} - u_{i-1}}{2h}, \quad u''(t_i) \approx \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2}.$$



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This yields a system of algebraic equations

$$\frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} = f\left(t_i, u_i, \frac{u_{i+1} - u_{i-1}}{2h}\right), \quad i = 1, \dots, n,$$

to be solved for the unknowns $u_i, i = 1, \dots, n$.

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In the matrix form, we have

$$\frac{1}{h^2} \begin{bmatrix} -2 & 1 & 0 & & \cdots & 0 \\ 1 & -2 & 1 & & & \\ \vdots & \vdots & \ddots & \ddots & & \vdots \\ 0 & \cdots & & 1 & -2 & 1 \\ & & & 0 & 1 & -2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} = \begin{bmatrix} f\left(t_1, u_1, \frac{u_2 - \alpha}{2h}\right) \\ f\left(t_2, u_2, \frac{u_3 - u_1}{2h}\right) \\ \vdots \\ f\left(t_n, u_n, \frac{\beta - u_{n-1}}{2h}\right) \end{bmatrix}$$

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which is denoted as

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Thus, the Newton's method for solving the system of algebraic equations is given by

$$u^{(m+1)} = u^{(m)} - \left[\frac{1}{h^2} A - F'(u^{(m)}) \right]^{-1} \left[\frac{1}{h^2} Au^{(m)} - F(u^{(m)}) - g \right]$$

where the Jacobian matrix is given by $[F(u)]_{ij} = [\partial f(t_i, u_i, (u_{i+1} - u_{i-1})/(2h))/\partial u_j]$.

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In particular,

$$\begin{aligned} [F'(u)]_{ii} &= f_2 \left(t_i, u_i, \frac{u_{i+1} - u_{i-1}}{2h} \right), \quad 1 \leq i \leq n, \\ [F'(u)]_{i,i-1} &= -\frac{1}{2h} f_3 \left(t_i, u_i, \frac{u_{i+1} - u_{i-1}}{2h} \right), \quad 2 \leq i \leq n, \\ [F'(u)]_{i,i+1} &= \frac{1}{2h} f_3 \left(t_i, u_i, \frac{u_{i+1} - u_{i-1}}{2h} \right), \quad 1 \leq i \leq n-1, \end{aligned}$$

where all other entries are 0 and $f_2(t, u, v)$, $f_3(t, u, v)$ denote the partial derivatives of f with respect to u and v respectively.

Example

Consider the two-point BVP

$$u'' = -u + \frac{2(u')^2}{u}, \quad -1 < t < 1,$$

$$u(-1) = u(1) = (e + e^{-1})^{-1}.$$

The iterative solution via Newton's method satisfies

$$u^{(m+1)} = u^{(m)} - \left[\frac{1}{h^2} A - F'(u^{(m)}) \right]^{-1} \left[\frac{1}{h^2} A u^{(m)} - F(u^{(m)}) - g \right],$$

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$$g = \frac{-1}{h^2} \begin{bmatrix} (e + e^{-1})^{-1} \\ 0 \\ \vdots \\ 0 \\ (e + e^{-1})^{-1} \end{bmatrix}.$$

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- Error Analysis



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We define

$$D_h^2 v(t) = \frac{v(t+h) - 2v(t) + v(t-h)}{h^2}.$$

Then, the local (truncation) error

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Now, if $u \in C^4([a, b])$, then

$$u(t_i + h) = u(t_i) + hu'(t_i) + \frac{h^2}{2}u''(t_i) + \frac{h^3}{6}u'''(t_i) + \frac{h^4}{24}u^{(4)}(\xi_1), \quad \xi_1 \in (t_i, t_i + h),$$

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Now, if $u \in C^4([a, b])$, then

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Thus,

$$\ell(t_i, u) = \frac{h^2}{24} \left(u^{(4)}(\xi_1) + u^{(4)}(\xi_2) \right)$$

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Now, if $u \in C^4([a, b])$, then

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Thus,

$$\ell(t_i, u) = \frac{h^2}{24} \left(u^{(4)}(\xi_1) + u^{(4)}(\xi_2) \right) = \frac{h^2}{12} u^{(4)}(\xi), \quad \xi \in (t_i - h, t_i + h).$$

Theorem

If $v \in C^2([a, b])$, then

$$\lim_{h \rightarrow 0} \|D_h^2 v - v''\|_{\infty, h} = 0.$$

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Proof:

We have already seen the proof for the second part. For the first part, we use

$$v(t_i + h) = v(t_i) + hv'(t_i) + \frac{h^2}{2} v''(\xi_1), \quad \xi_1 \in (t_i, t_i + h),$$

$$v(t_i - h) = v(t_i) - hv'(t_i) + \frac{h^2}{2} v''(\xi_2), \quad \xi_2 \in (t_i - h, t_i),$$

yielding

$$D_h^2 v(t_i) - v''(t_i) = \frac{v''(\xi_1) + v''(\xi_2)}{2} - v''(t_i) = v''(\xi) - v''(t_i), \quad \xi \in (t_i - h, t_i + h).$$

The result follows!



Theorem (Discrete Maximum Principle)

Let v be a function on $[a, b]$ satisfying $D_h^2 v \geq 0$ on $t_i, i = 1, \dots, n$. Then $\max_{1 \leq i \leq n} v(t_i) \leq \max\{v(t_0), v(t_{n+1})\}$. Equality holds if and only if v is constant.



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There is a unique solution to the discrete BVP

$$D_h^2 u_h(t_i) = f(t_i), \quad t_i, i = 1, \dots, n, \\ u_h(a) = \alpha, \quad u_h(b) = \beta.$$



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The quantity $\|f - D_h^2 u\| = \|u'' - D_h^2 u\|$ is the **consistency error** of the discretization and the statement $\lim_{h \rightarrow 0} \|u'' - D_h^2 u\| = 0$ means that the discretization is **consistent**.



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As a final remark, the finite difference method helped us find the solution values at the mesh points, but the solution at non-mesh points are not readily available from the method. If needed, one can obtain the solution at non-mesh points through interpolation or try other approximation approaches ...