

Lecture 24

Last time we studied iterative method to solve system of linear equations

$$Ax = b$$

The method used is based on the idea of an approximate inverse to A , i.e., a matrix C such that

$$\|I - CA\| < 1 \quad \text{for some matrix norm}$$

The iteration function g is given by

$$\begin{aligned} g(\bar{x}) &= C\bar{b} + (I - CA)\bar{x} \\ &= \bar{x} + C(\bar{b} - A\bar{x}) \end{aligned}$$

Thus the iteration is

x_0 any

$$x^{(m+1)} = x^{(m)} + C(b - Ax^{(m)})$$

for $m = 0, 1, 2, \dots$

Then $x^{(m+1)} \rightarrow \xi$ fixed pt of g
also $A\xi = b$.

Two of the most common methods are
the . Jacobi-method
• Gauss-Siedel method.

Gauss-Siedel method is usually faster than
Jacobi-method.

Gauss-Siedel method converges for
strictly row-dominant matrices and positive
definite matrices.

Last time we proved that the Jacobi method
converges for strictly row-dominant matrices.

Iterative methods are usually used when
we have large system of linear equations with
a sparse coefficient matrix.

For the next few lectures we will be learning numerical methods to solve initial value problems.

$$\frac{dy}{dx} = f(x, y)$$

$$x_0 \leq x \leq x_n$$

$$y(x_0) = y_0$$

We assume that all partial derivatives to f exist.

By theory there exists a unique solution to the above initial value problem (in a neighbourhood of (x_0, y_0)).

However, one cannot hope for exact solutions

Example

$$\frac{dy}{dx} = \sin(x^2 + y^2)$$

$$y(0) = 1$$

$$0 \leq x \leq 1$$

We cannot find exact answer to above equation

Numerical integration by Taylor Series

$$(*) \quad \frac{dy}{dx} = f(x, y) \quad y(x_0) = y_0$$

If $y(x)$ is exact solution of $(*)$ then we can expand $y(x)$ into a Taylor series about the pt x_0

$$(**) \quad y(x) = y_0 + (x-x_0) y'(x_0) + \frac{(x-x_0)^2}{2!} y''(x_0) + \dots$$

The derivatives in this expansion are not known explicitly since the solution is unknown. However if f is sufficiently differentiable they can be obtained by taking the total derivative of $(*)$ with respect to x , keeping in mind that y itself is a function of x .

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

$$y'' = f' = f_x + f_y f$$

$$y''' = f'' = f_{xx} + f_{xy}f + f_{yx}f + f_{yy}f^2 + f_{yx}f + f_{yy}f^2 + f_{yy}f^2$$

Continuing in this manner, we can express any derivative of y in terms of f and its partial derivatives. It is already clear, however, that unless f is a very simple function, the higher total derivatives become increasingly complex.

For practical reasons then, one must limit the number of terms in the expansion (**) to a reasonable number.

If we assume that the truncated series (**) is a good approximation for a step of length h , i.e., for $x - x_0 = h$, we can then evaluate y at $x_0 + h$, reevaluate the derivatives y' , y'' etc. and then use (**) to

proceed to the next step.

Thus we obtain a discrete set of values y_n which are approximations to the true solution at the points $x_n = x_0 + nh$ ($n = 0, 1, 2, \dots$).

Remark We will always denote the value of the exact solution at a pt x_n by $y(x_n)$ and of an approximate solution by y_n .

define

$$T_k(x, y) = f(x, y) + \frac{h}{2!} f'(x, y) + \frac{h^2}{3!} f''(x, y) + \dots + \frac{h^{k-1}}{k!} f^{(k-1)}(x, y)$$

$$k = 1, 2, \dots$$

Algorithm 1 (Taylor algorithm of order k)

To find an approximate solution of the diff equation

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

$$y(a) = y_0$$

over an interval $[a, b]$

1. Choose a step $h = \frac{b-a}{N}$.

$$\text{Set } x_n = a + nh \quad n = 0, 1, \dots, N$$

2. Generate approximations y_n to $y(x_n)$ from the recursion

$$y_{n+1} = y_n + h T_k(x_n, y_n) \quad n = 0, 1, \dots, N-1$$

Taylor's theorem with remainder shows that the local error of Taylor's algorithm of order k

$$\text{is } E = \frac{h^{(k+1)} f^{(k)}(\xi, y(\xi))}{(k+1)!} \quad x_n < \xi < x_n + h$$

$$E = \frac{h^{k+1}}{(k+1)!} y^{(k+1)}(\xi)$$

Example Use Taylor's method of order 2 to solve

$$\frac{dy}{dx} = y - x^2 + 1 \quad 0 \leq x \leq 2$$

$$y(0) = 0.5$$

$$\begin{aligned} f' &= \frac{d}{dx} (y - x^2 + 1) \\ &= \frac{dy}{dx} - 2x \\ &= y - x^2 + 1 - 2x \end{aligned}$$

$$\begin{aligned} T_2(x_i, y_i) &= f(x_i, y_i) + \frac{h}{2} f'(x_i, y_i) \\ &= y_i - x_i^2 + 1 + \frac{h}{2} (y_i - x_i^2 - 2x_i + 1) \\ &= \left(1 + \frac{h}{2}\right) (y_i - x_i^2 + 1) + \frac{h}{2} \end{aligned}$$

Taylor method of order 2

$$y_0 = 0.5$$

$$y_{i+1} = y_i + h \left[\left(1 + \frac{h}{2}\right) (y_i - x_i^2 + 1) - h x_i \right]$$

Exact solution to above d.e is

$$y(x) = (x+1)^2 - 0.5e^x$$

x_i	Exact $y(x_i)$	Taylor order 2 y_i	Error
0.0	0.5	0.5	0
0.2	0.8292986	0.8300000	0.0007014
0.4	1.2140877	1.2158000	0.0017123
0.6	1.6489406	1.6520760	0.0031354
0.8	2.1272295	2.1323327	0.0051032
1.0	2.6408591	2.6486459	0.0077868
1.2			
1.4			
1.6			
1.8			
2.0	5.3054720	5.3476843	0.0422123

On setting $k=1$ in our Algorithm we obtain Euler's method

$$y_{n+1} = y_n + h f(x_n, y_n)$$

local error $E = \frac{h^3}{2} y''(\xi)$
 $x_n \leq \xi \leq x_{n+1}$

Example

$$\frac{dy}{dx} = y - x^2 + 1$$

$$0 \leq x \leq 2$$

$$y(0) = 0.5$$

$$h = 0.2$$

$$x_n = 0.2n$$

$$y_{n+1} = y_n + 0.2 (y_n - 0.04n^2 + 1)$$

Exact solⁿ $y(x) = (x+1)^2 - 0.5e^x$

x_i	Euler method y_i	Exact $y(x_i)$	Error $ y_i - y(x_i) $
0	0.5	0.5	0
0.2	0.8	0.8292986	0.0292981
0.4	1.152	1.2140872	0.0620872
0.6	1.5504	1.6489406	0.0985406
0.8	1.9884800	2.1272295	0.1387495
1.0	2.4581760	2.6408591	0.1826831
1.2			
1.4			
1.6			
1.8			
2.0	4.8657845	5.3054720	0.4396854

Notice error grows slightly as the value of x increases.

This controlled error growth is a consequence of the stability of Euler's method, which implies that the error is expected to grow in no worse than a linear manner.

Error estimates and convergence of Euler's method

To solve the d-e

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

$$y(x_0) = y_0$$

by Euler's method, we choose a constant step size h and we apply the formula

$$(*) \quad y_{n+1} = y_n + h f(x_n, y_n) \\ n = 0, 1, 2, \dots$$

where $x_n = x_0 + nh$.

We denote the true solution of the d-e. at $x = x_n$ by $y(x_n)$ and the approximate solⁿ obtained by applying (*) as y_n .

We wish to estimate the magnitude of the discretization error e_n defined by

$$e_n = y(x_n) - y_n$$

Assuming that appropriate derivatives exist, we can expand $y(x_{n+1})$ about $x = x_n$ using Taylor's theorem with remainder

$$(xx) \quad y(x_{n+1}) = y(x_n) + h y'(x_n) + \frac{h^2}{2} y''(\xi_n)$$

$$x_n \leq \xi \leq x_{n+1}$$

$$E_n = \frac{h^2}{2} y''(\xi_n) \rightarrow \text{local discretization error}$$

This is the error committed in the single step from x_n to x_{n+1} assuming that y and y' are known exactly at $x = x_n$

(we ignore Round-off error in this section)

On subtracting (*) from (**) we obtain

$$e_{n+1} = e_n + h \left[f(x_n, y(x_n)) - f(x_n, y_n) \right] + \frac{h^2}{2} y''(\xi_n)$$

By M.V.T

$$\begin{aligned} f(x_n, y(x_n)) - f(x_n, \bar{y}_n) &= \frac{\partial f}{\partial y}(x_n, \bar{y}_n) (y(x_n) - \bar{y}_n) \\ &= f_y(x_n, \bar{y}_n) \cdot e_n \end{aligned}$$

where \bar{y}_n is between \bar{y}_n and $y(x_n)$

So we have

$$e_{n+1} = e_n + h f_y(x_n, \bar{y}_n) e_n + \frac{h^2}{2} y''(\xi_n)$$

Theorem Let y_n be the approximate solⁿ to

$$\frac{dy}{dx} = f(x, y) \quad y(x_0) = y_0$$

If the exact solⁿ $y(x)$ of above d.e has a continuous second derivative on the interval $[x_0, b]$ and if on this interval

the inequalities

$$|f_y(x, y)| \leq L, \quad |y''(x)| < \gamma$$

are satisfied for fixed the constants L and γ , the error $e_n = y(x_n) - y_n$ of Euler's method at a pt $x_n = x_0 + nh$ is bounded as follows

$$|e_n| \leq \frac{h\gamma}{2L} (e^{(x_n - x_0)L} - 1)$$

Pf

$$e_{n+1} = e_n + hf_y(x_n, \bar{y}_n)e_n + \frac{h^2}{2} y''(\xi_n)$$

$$\begin{aligned} |e_{n+1}| &\leq |e_n| + hL|e_n| + \frac{h^2}{2}\gamma \\ &= (1+hL)|e_n| + \frac{h^2}{2}\gamma \end{aligned}$$

We show by induction that the solution of the difference equation

$$\xi_{n+1} = (1+hL)\xi_n + \frac{h^2}{2}\gamma, \quad \xi_0 = 0$$

dominates $|e_n|$, i.e we will show

$$(*) \quad \xi_n \geq |e_n| \quad \text{for } n=0,1,-\dots$$

Since $e_0 = \xi_0 = 0$, $(*)$ is true for $n=0$.

assume true of $(*)$ for n

$$\xi_n \geq |e_n|$$

$$\xi_{n+1} = (1+h\gamma)\xi_n + \frac{h^2}{2}\gamma$$

$$\geq (1+h\gamma)|e_n| + \frac{h^2}{2}\gamma$$

$$\geq |e_{n+1}|$$

Thus $\xi_n \geq |e_n| \quad \forall n=0,1,2,-\dots$

$$\begin{cases} \xi_{n+1} = (1+h\gamma)\xi_n + \frac{h^2}{2}\gamma \\ \xi_0 = 0 \end{cases}$$

solⁿ $\xi_n = c(1+hL)^n - B$

where $B = \frac{h\gamma}{2L}$ and c is a constant to be determined

$$\xi_0 = 0$$

$$\text{so } c = B$$

$$\xi_n = B(1+hL)^n - B$$

$$e^x = 1+x + e^{\frac{x}{2}} \frac{x^2}{2}$$

Thus $e^x \geq 1+x$ for all x

$$1+hL \leq e^{hL}$$

$$\text{so } (1+hL)^n \leq e^{nhL}$$

$$\text{Therefore } \xi_n \leq B(e^{nhL} - 1)$$

$$= \frac{h\gamma}{2L} (e^{nhL} - 1)$$

$$= \frac{h\gamma}{2L} (e^{(x_n - x_0)L} - 1)$$

$$\text{note } nh = x_n - x_0$$

Since $|e_n| \leq \varepsilon_n$

we have proved the theorem.