13. NUMERICAL SOLUTION OF ORDINARY DIFFERENTIAL EQUATIONS: BACKGROUND

Consider the initial value problem (IVP) for a first order ordinary differential equation:

$$dy/dx = f(x, y), \qquad y(x_0) = y_0.$$

The following theorem gives sufficient conditions for existence and uniqueness of a solution.

Theorem 10. Let f(x, y) satisfy the following conditions:

(A) f(x,y) is defined and continuous in the strip $x_0 \leq x \leq b, -\infty < y < \infty$, where x_0 and b are finite.

(B) There exists a constant L such that for any $x \in [x_0, b]$ and any two numbers y and y^* , $|f(x, y) - f(x, y^*)| \le L|y - y^*|$.

Then given any number y_0 , there exists exactly one function y(x) satisfying: (i) y(x) is continuous and differentiable on $[x_0, b]$, (ii) y'(x) = f(x, y(x)), $x \in [x_0, b]$, and (iii) $y(x_0) = y_0$, i.e., the IVP has a unique solution.

It is also possible to view y as a vector with N components, so that the IVP represents a first order system of odes. One way to treat higher order odes is to reduce them to a first order system by introducing additional variables:

Example: $d^2y/dx^2 = f(x, y, dy/dx)$. Set z = dy/dx. Then dz/dx = f(x, y, z) and we obtain the first order system:

$$\frac{d}{dx}\begin{pmatrix} y\\z \end{pmatrix} = \begin{pmatrix} z\\f(x,y,z) \end{pmatrix} = \begin{pmatrix} f_1(x,y,z)\\f_2(x,y,z) \end{pmatrix}.$$

13.1. Euler's method. Our numerical schemes will seek approximations to the solution y(x) at a sequence of points x_i , i.e., we will approximate $y(x_i)$ by a number y_i . We begin by discussing the simplest method, i.e., Euler's method. Set $y_0 = y(x_0)$ and define

$$y_{n+1} = y_n + h_n f(x_n, y_n), \qquad n = 0, 1, \dots,$$

where $h_n = x_{n+1} - x_n$.

One motivation of this method is that we have approximated the derivative $(dy/dx)(x_n)$ by the forward difference approximation $(y(x_{n+1}) - y(x_n))/(x_{n+1} - x_n)$ and so:

$$y(x_{n+1}) \approx y(x_n) + h_n f(x_n, y(x_n)).$$

We then define our approximations y_n as the value that restores equality, i.e., $y_{n+1} = y_n + h_n f(x_n, y_n)$.

Another motivation for the method is to expand the solution in a Taylor series expansion and neglect higher order terms, i.e.,

$$y(x_n + h_n) = y(x_n) + h_n y'(x_n) + O(h_n^2)$$

= $y(x_n) + h_n f(x_n, y(x_n)) + O(h_n^2) \approx y(x_n) + h_n f(x_n, y(x_n)).$

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Example: y' = y y(0) = 1. Then Euler's method, using a constant step size $h_n = h$, is: $y_{n+1} = y_n + hy_n = (1+h)y_n$. Hence $y_0 = 1$, $y_1 = 1 + h$, $y_2 = (1+h)y_1 = (1+h)^2$, and $y_n = (1+h)^n$.

We next consider the convergence of Euler's method. Expanding the solution y(x) in a Taylor series, we have

$$y(x_{n+1}) = y(x_n) + h_n f(x_n, y(x_n)) + (h_n^2/2)y''(\xi_n), \qquad x_n \le \xi_n \le x_{n+1}.$$

Neglecting any roundoff errors, the approximation given by Euler's method satisfies:

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

Let $e_n = y(x_n) - y_n$. Note $e_0 = 0$. Subtracting equations, we get

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

Hence

$$|e_{n+1}| \le |e_n| + h_n |f(x_n, y(x_n)) - f(x_n, y_n)| + (h_n^2/2) |y''(\xi_n)|$$

$$\le |e_n| + h_n L e_n + h_n^2 M_2/2 \le (1 + h_n L) |e_n| + h_n^2 M_2/2,$$

where we assume that $\max |y''(x)| \leq M_2$. Consider the case when $h_n = h$ for all n. Then

$$\begin{aligned} |e_1| &\leq h^2 M_2/2, \qquad |e_2| \leq (1+hL)|e_1| + h^2 M_2/2 \leq [1+(1+hL)]h^2 M_2/2, \\ |e_3| &\leq (1+hL)|e_2| + h^2 M_2/2 \leq [1+(1+hL)+(1+hL)^2]h^2 M_2/2. \end{aligned}$$

Using the fact that $\sum_{i=0}^{n-1} r^i = (1-r^n)/(1-r)$, we get

$$|e_n| \le [1 + (1 + hL) + (1 + hL)^2 + \dots + (1 + hL)^{n-1}]h^2M_2/2 \le [(1 + hL)^n - 1]hM_2/(2L).$$

Observing that $e^x = 1 + x + e^{\xi} x^2/2 \ge 1 + x$ for all x, we see that $1 + hL \le e^{hL}$ and hence $(1 + hL)^n \le e^{nhL} = e^{(x_n - x_0)L}$. Thus, we get the error estimate:

$$|e_n| \le \frac{hM_2}{2L} [e^{(x_n - x_0)L} - 1],$$

so the error bound is O(h). This bound is quite pessimistic and not a realistic way to determine a value of h to guarantee a given accuracy. It also requires a bound on y''.

We now consider what this result says about convergence of Euler's method, and first what we mean by convergence in this context.

Let x be a point in the interval $[x_0, b]$ and let y(x) denote the true solution of the IVP at the point x. For each value of the step size h, we will have an approximation to y(x) that we denote by y_n^h , where n will be determined by the equation $x - x_0 = nh$. Thus, as h is decreased, the value of n for which y_n denotes the approximation to y(x) will also change. So for convergence, we want:

$$\lim_{\substack{h \to 0 \\ n \to \infty \\ nh = x}} y_n^h = y(x).$$

Example: For $x_0 = 0$, x = 1/2, and the sequence h = 1/4, 1/8, 1/16, 1/32, we look for the convergence of $y_2^{1/4}, y_4^{1/8}, y_8^{1/16}, y_{16}^{1/32}$.

Suppose in the error estimate for Euler's method, we keep $x_n = x$ fixed, i.e., we choose n so that $nh = x - x_0$ and let $h \to 0$. Then

$$|y(x) - y_n^h| \le \frac{hM_2}{2L} [e^{(x-x_0)L} - 1] \implies \lim_{\substack{h \to 0 \\ n \to \infty \\ nh = x}} |y(x) - y_n^h| = 0,$$

so we have converence of the method as $h \to 0$.

13.2. Taylor series methods. Consider the Taylor series of y(x), the solution of the IVP, about $x = x_n$:

$$y(x_n + h) = y(x_n) + hy'(x_n) + \frac{h^2}{2}y^{(2)}(x_n) + \dots + \frac{h^k}{k!}y^{(k)}(x_n) + \frac{h^{k+1}}{(k+1)!}y^{(k+1)}(\xi)$$

Now y'(x) = f(x, y(x)), so

$$y''(x) = f'(x, y(x)) = \frac{d}{dx}f(x, y(x)) = f_x(x, y(x)) + f_y(x, y(x))\frac{dy}{dx}$$

In general,

$$y^{(k)}(x) = f^{(k-1)}(x, y(x)) = \frac{d}{dx} f^{(k-2)}(x, y(x)) = f_x^{(k-2)}(x, y(x)) + f_y^{(k-2)}(x, y(x)) \frac{dy}{dx}.$$

Hence, if $y(x_n)$ were known, we could compute an approximation to $y(x_n + h)$ by using the truncated Taylor series:

$$y(x_n+h) \approx y(x_n) + hf(x_n, y(x_n)) + \frac{h^2}{2}f'(x_n, y(x_n)) + \dots + \frac{h^k}{k!}f^{(k-1)}(x_n, y(x_n)),$$

i.e., if we denote by y_n the approximation to $y(x_n)$, we can define the Taylor algorithm of order k as the sequence of computations

$$y_{n+1} = y_n + hT_k(x_n, y_n), \qquad n = 0, 1, \dots,$$

where

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

Note: Euler's method is the Taylor algorithm of order 1.

Example: We wish to solve the IVP $y' = 1/x^3 - y/x - y^2$, y(1) = 2 by the Taylor algorithm of order 2. Now

$$f(x,y) = \frac{1}{x^3} - \frac{y}{x} - y^2$$

and

$$\begin{aligned} f'(x,y) &= f_x(x,y) + f_y(x,y)f(x,y) = -\frac{3}{x^4} + \frac{y}{x^2} - \left[\frac{1}{x} + 2y\right]f\\ &= -\frac{3}{x^4} + \frac{y}{x^2} - \left[\frac{1}{x} + 2y\right]\left[\frac{1}{x^2} - \frac{y}{x} - y^2\right]. \end{aligned}$$

Then

$$y_{n+1} = y_n + hf(x_n.y_n) + \frac{h^2}{2}f'(x_n,y_n),$$

where f'(x, y) defined above.

Note we can also compute f'(x, y) directly, remembering that y is a function of x, i.e.,

$$\begin{aligned} f'(x,y) &= \frac{d}{dx} \left[\frac{1}{x^3} - \frac{y}{x} - y^2 \right] = -\frac{3}{x^4} - \frac{xy' - y}{x^2} - 2yy' = -\frac{3}{x^4} + \frac{y}{x^2} - \left[\frac{1}{x} + 2y \right] y' \\ &= -\frac{3}{x^4} + \frac{y}{x^2} - \left[\frac{1}{x} + 2y \right] \left[\frac{1}{x^2} - \frac{y}{x} - y^2 \right]. \end{aligned}$$

Definition: The local truncation error for the Taylor series method of order k is defined by:

$$y(x_{n+1}) - y(x_n) - hT_k(x_n, y(x_n)) = \frac{h^{k+1}}{(k+1)!}y^{(k+1)}(\xi_n)$$

The local truncation of Euler's method is $h^2 y^{(2)}(\xi_n)/2$.

The Taylor algorithm of order k is an example of a one-step method, i.e, the value of y_{n+1} only depends on one past value (y_n) . One-step methods have the form

$$y_{n+1} = y_n + h\Phi(x_n, y_n), \qquad n = 0, 1, \dots,$$

Analogously to the Taylor series methods, we define the Local Truncation Error of such methods to be

$$LTE = y(x_{n+1}) - y(x_n) - h\Phi(x_n, y(x_n))$$

Then we have the following result giving a bound on the global error.

Theorem 11. If $|\Phi(x, u) - \Phi(x, v)| \leq \mathcal{L}|u - v|$ for $a \leq x \leq b$, $0 < h < h_0$ and all u, v and if the local trucation error is $O(h^{p+1})$, then for any $x_n = x_0 + nh \in [x_0, b]$,

$$|y(x_n) - y_n| \le C \frac{h^p}{\mathcal{L}} (e^{\mathcal{L}(x_n - x_0)} - 1).$$

The proof of this result is essentially identical to the proof of the error bound for Euler's method.

Although Taylor series methods become increasingly more accurate as k increases, their major disadvantage is that they require calulation of high derivatives of the function f.