Queens College, CUNY, Department of Computer Science Numerical Methods CSCI 361 / 761 Spring 2018

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17 Lecture 17

Numerical solution of systems of ordinary differential equations

- In this lecture we continue the study of initial value problems.
- This lecture is devoted to formal mathematical proofs.
- The contents of this lecture are not for examination.

17.1 Basic notation

- We repeat the basic definitions from the previous lecture.
- Let the system of coupled differential equations be

$$\frac{d\mathbf{y}}{dx} = \mathbf{f}(x, \mathbf{y}). \tag{17.1.1}$$

- There are m unknown variables $y_j, j = 1, \ldots, m$.
- The starting point is $x = x_0$, and the initial value y_0 is given.
- The above is called the Cauchy problem or initial value problem.
- Our interest is to integrate eq. (17.1.1) numerically, using steps h_i , so $x_{i+1} = x_i + h_i$.
- The steps h_i need not be of equal size.
- We define $y_i = y(x_i)$.

17.2 Trapezoid and midpoint methods

- The trapezoid and midpoint methods are both explicit second order numerical integration algorithms.
- Recall the formula for the trapezoid method:

$$\mathbf{g}_1 = \mathbf{f}(x_i, \mathbf{y}_i), \qquad (17.2.1a)$$

$$g_2 = f(x_i + h_i, y_i + h_i g_1),$$
 (17.2.1b)

$$\mathbf{y}_{i+1} = \mathbf{y}_i + \frac{h_i}{2} (\mathbf{g}_1 + \mathbf{g}_2).$$
 (17.2.1c)

• Recall the formula for the midpoiint method:

$$\mathbf{g}_1 = \mathbf{f}(x_i, \mathbf{y}_i), \tag{17.2.2a}$$

$$\mathbf{g}_2 = \mathbf{f}(x_i + \frac{1}{2}h_i, \mathbf{y}_i + \frac{1}{2}h_i \mathbf{g}_1), \qquad (17.2.2b)$$

$$y_{i+1} = y_i + h_i g_2. (17.2.2c)$$

- Because the trapezoid and midpoint methods are second order integration methods, in both cases the numerical error in one integration step is of $O(h_i^3)$.
- In this lecture we shall give formal proofs that the numerical error in one integration step of the trapezoid and midpoint methods is of $O(h_i^3)$.

17.3 Derivation of numerical error of trapezoid method

- To keep the proof simple, we shall treat only the case of one unknown, which we shall call y.
- The differential equation is therefore

$$\frac{dy}{dx} = f(x,y). ag{17.3.1}$$

- Since we are analyzing only one integration step, we denote the stepsize by h.
- The formula for the trapezoid method is

$$y_{i+1} = y_i + \frac{h}{2} \left[f(x_i, y_i) + f(x_i + h, y_i + hf(x_i, y_i)) \right].$$
 (17.3.2)

• The Taylor series for y(x) is

$$y(x_i + h) = y(x_i) + h\frac{dy}{dx} + \frac{h^2}{2!}\frac{d^2y}{dx^2} + \frac{h^3}{3!}\frac{d^3y}{dx^3} + \cdots$$
 (17.3.3)

- All the derivatives in eq. (17.3.3) are evaluated at $x = x_i$.
- Recall that for the explicit Euler method, we substituted for dy/dx to obtain

$$y(x_i + h) = y(x_i) + h f(x_i, y_i) + \frac{h^2}{2!} \frac{d^2 y}{dx^2} + \cdots$$
 (17.3.4)

- This was the proof that the error term for the explicit Euler method is $O(h^2)$, hence it is a first order method.
- We need to do more work to process the trapezoid method.
- We differentiate eq. (17.3.1) to obtain

$$\frac{d^2y}{dx^2} = \frac{df(x,y)}{dx}
= \frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} \frac{dy}{dx}
= \frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} f.$$
(17.3.5)

 \bullet For brevity, define a parameter k via

$$k = h f(x_i, y_i)$$
. (17.3.6)

• Next perform a two-dimensional Taylor series for $f(x_i+h, y_i+k)$ as follows. All the derivatives (and partial derivatives) are evaluated at (x_i, y_i) . Then we obtain

$$f(x_{i} + h, y_{i} + k) = f(x_{i}, y_{i}) + h \frac{\partial f}{\partial x} + k \frac{\partial f}{\partial y}$$

$$+ \frac{h^{2}}{2} \frac{\partial^{2} f}{\partial x^{2}} + hk \frac{\partial^{2} f}{\partial x \partial y} + \frac{k^{2}}{2} \frac{\partial^{2} f}{\partial y^{2}} + \cdots$$

$$= f(x_{i}, y_{i}) + h \frac{\partial f}{\partial x} + h \frac{\partial f}{\partial y} f$$

$$+ \frac{h^{2}}{2} \frac{\partial^{2} f}{\partial x^{2}} + h^{2} \frac{\partial^{2} f}{\partial x \partial y} f + \frac{h^{2}}{2} \frac{\partial^{2} f}{\partial y^{2}} f^{2} + \cdots$$

$$(17.3.7)$$

• Substituting in eq. (17.3.2) yields

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

$$= y(x_i) + \frac{h}{2} \left[f(x_i, y_i) + f(x_i + h, y_i + k) \right]$$

$$= y(x_i) + h f(x_i, y_i) + \frac{h^2}{2} \left(\frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} f \right) + O(h^3)$$

$$= y(x_i) + h \frac{dy}{dx} + \frac{h^2}{2} \frac{d^2 y}{dx^2} + O(h^3) .$$
(17.3.8)

• Hence the error term is of order $O(h^3)$ and the integration method is of second order.

17.4 Derivation of numerical error of midpoint method

- The proof is very similar to that for the trapezoid method in Sec. 17.3.
- Once again, we treat only the case of one unknown, which we shall call y.
- The differential equation is therefore given by eq. (17.3.1).
- We again denote the integration step by h.
- The formula for the midpoint method is

$$y_{i+1} = y_i + \frac{h}{2} f(x_i + \frac{1}{2}h, y_i + \frac{1}{2}hf(x_i, y_i)).$$
 (17.4.1)

- The Taylor series for y(x) is given by eq. (17.3.3).
- Recall also

$$\frac{d^2y}{dx^2} = \frac{\partial f}{\partial x} + \frac{\partial f}{\partial y}f. \tag{17.4.2}$$

- For brevity, define a parameter k via $k = h f(x_i, y_i)$
- Next perform a two-dimensional Taylor series for $f(x_i + \frac{1}{2}h, y_i + \frac{1}{2}k)$ as follows. All the derivatives (and partial derivatives) are evaluated at (x_i, y_i) . Then we obtain

$$f(x_{i} + \frac{1}{2}h, y_{i} + \frac{1}{2}k) = f(x_{i}, y_{i}) + \frac{h}{2} \frac{\partial f}{\partial x} + \frac{k}{2} \frac{\partial f}{\partial y}$$

$$+ \frac{h^{2}}{8} \frac{\partial^{2} f}{\partial x^{2}} + \frac{hk}{4} \frac{\partial^{2} f}{\partial x \partial y} + \frac{k^{2}}{8} \frac{\partial^{2} f}{\partial y^{2}} + \cdots$$

$$= f(x_{i}, y_{i}) + \frac{h}{2} \frac{\partial f}{\partial x} + \frac{h}{2} \frac{\partial f}{\partial y} f$$

$$+ \frac{h^{2}}{8} \frac{\partial^{2} f}{\partial x^{2}} + \frac{h^{2}}{4} \frac{\partial^{2} f}{\partial x \partial y} f + \frac{h^{2}}{8} \frac{\partial^{2} f}{\partial y^{2}} f^{2} + \cdots$$

$$(17.4.3)$$

• Substituting in eq. (17.4.1) yields

$$y_{i+1} = y_i + h f(x_i + \frac{1}{2}h, y_i + \frac{1}{2}hf(x_i, y_i))$$

$$= y(x_i) + h f(x_i + \frac{1}{2}h, y_i + \frac{1}{2}k)$$

$$= y(x_i) + h f(x_i, y_i) + \frac{h^2}{2} \left(\frac{\partial f}{\partial x} + \frac{\partial f}{\partial y}f\right) + O(h^3)$$

$$= y(x_i) + h \frac{dy}{dx} + \frac{h^2}{2} \frac{d^2y}{dx^2} + O(h^3).$$
(17.4.4)

• Hence the error term is of order $O(h^3)$ and the integration method is of second order.

17.5 Lessons to learn

- There are various lessons to learn from the proofs in Secs. 17.3 and 17.4.
- In both cases, the proofs were based on Taylor series expansions.
- The above fact is significant.
 - 1. In both proofs we assumed that y(x) is twice differentiable, i.e. that d^2y/dx^2 exists.
 - 2. This in turn assumes that f(x,y) is differentiable in both x and y, i.e. that $\partial f/\partial x$ and $\partial f/\partial y$ both exist.
 - 3. In the case of m > 1 unknowns y_1, \ldots, y_m and m right-hand side functions $f_k(x, y)$, $k = 1, \ldots, n$, we require the **Jacobian matrix** to exist. This is the matrix of the m^2 partial derivatives $\partial f_k/\partial y_i$, $1 \le j, k \le m$.
- However, it is possible that y(x) is not twice differentiable. It is possible that f is continuous but does not have well-defined partial derivatives for all m^2 entries in the Jacobian matrix.
- The proofs for higher order integration methods, such as fourth order Runge–Kutta RK4, are all based on Taylor series expansions.
- The higher the order of an integration method, the higher also the order of the partial derivatives of f which are required to exist, to justify the proof.
- This leads to the warning that higher order does not always imply higher accuracy.
- If y(x) is not twice differentiable, for example, a second (or higher) order integration method will not necessarily yield good results, or in any case it may not perform better than a first order integration method.
- The fourth order Runge–Kutta method RK4 is widely used and is very popular. It is simple to code and works well in a wide variety of practical applications. Going beyond fourth order generally does not yield much additional benefit, relative to the extra computations involved.
- Hence most people do not employ integration methods beyond the fourth order.