

Queens College, CUNY, Department of Computer Science

Numerical Methods

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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

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

- There are n unknown variables y_j , $j = 1, \dots, n$.
- The starting point is $x = x_0$, and the initial value \mathbf{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 $\mathbf{y}_i = \mathbf{y}(x_i)$.

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:

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

$$\mathbf{g}_2 = \mathbf{f}(x_i + h_i, \mathbf{y}_i + h_i \mathbf{g}_1), \quad (17.2.1b)$$

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

- Recall the formula for the midpoint method:

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

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

$$\mathbf{y}_{i+1} = \mathbf{y}_i + h_i \mathbf{g}_2. \quad (17.2.2c)$$

- 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

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

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

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

- The Taylor series for $y(x)$ is

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

- 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

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

- 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

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

- For brevity, define a parameter k via

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

- 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

$$\begin{aligned}
f(x_i+h, y_i+k) &= f(x_i, y_i) + h \frac{\partial f}{\partial x} + k \frac{\partial f}{\partial y} \\
&\quad + \frac{h^2}{2} \frac{\partial^2 f}{\partial x^2} + hk \frac{\partial^2 f}{\partial x \partial y} + \frac{k^2}{2} \frac{\partial^2 f}{\partial y^2} + \dots \\
&= f(x_i, y_i) + h \frac{\partial f}{\partial x} + h \frac{\partial f}{\partial y} f \\
&\quad + \frac{h^2}{2} \frac{\partial^2 f}{\partial x^2} + h^2 \frac{\partial^2 f}{\partial x \partial y} f + \frac{h^2}{2} \frac{\partial^2 f}{\partial y^2} f^2 + \dots
\end{aligned} \tag{17.3.7}$$

- Substituting in eq. (17.3.2) yields

$$\begin{aligned}
y_{i+1} &= y_i + \frac{h}{2} \left[f(x_i, y_i) + f(x_i+h, y_i+h f(x_i, y_i)) \right] \\
&= y(x_i) + \frac{h}{2} \left[f(x_i, y_i) + f(x_i+h, y_i+k) \right] \\
&= y(x_i) + h f(x_i, y_i) + \frac{h^2}{2} \left(\frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} f \right) + O(h^3) \\
&= y(x_i) + h \frac{dy}{dx} + \frac{h^2}{2} \frac{d^2 y}{dx^2} + O(h^3).
\end{aligned} \tag{17.3.8}$$

- 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

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

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

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

- For brevity, define a parameter k via $k = hf(x_i, y_i)$.
- 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

$$\begin{aligned} f\left(x_i + \frac{1}{2}h, y_i + \frac{1}{2}k\right) &= f(x_i, y_i) + \frac{h}{2} \frac{\partial f}{\partial x} + \frac{k}{2} \frac{\partial f}{\partial y} \\ &\quad + \frac{h^2}{8} \frac{\partial^2 f}{\partial x^2} + \frac{hk}{4} \frac{\partial^2 f}{\partial x \partial y} + \frac{k^2}{8} \frac{\partial^2 f}{\partial y^2} + \dots \\ &= f(x_i, y_i) + \frac{h}{2} \frac{\partial f}{\partial x} + \frac{h}{2} \frac{\partial f}{\partial y} f \\ &\quad + \frac{h^2}{8} \frac{\partial^2 f}{\partial x^2} + \frac{h^2}{4} \frac{\partial^2 f}{\partial x \partial y} f + \frac{h^2}{8} \frac{\partial^2 f}{\partial y^2} f^2 + \dots \end{aligned} \quad (17.4.3)$$

- Substituting in eq. (17.4.1) yields

$$\begin{aligned} y_{i+1} &= y_i + hf\left(x_i + \frac{1}{2}h, y_i + \frac{1}{2}hf(x_i, y_i)\right) \\ &= y(x_i) + hf\left(x_i + \frac{1}{2}h, y_i + \frac{1}{2}k\right) \\ &= y(x_i) + hf(x_i, y_i) + \frac{h^2}{2} \left(\frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} f \right) + O(h^3) \\ &= y(x_i) + h \frac{dy}{dx} + \frac{h^2}{2} \frac{d^2y}{dx^2} + O(h^3). \end{aligned} \quad (17.4.4)$$

- 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 $n > 1$ unknowns y_1, \dots, y_n and n right-hand side functions $f_m(x, \mathbf{y})$, $m = 1, \dots, n$, we require the **Jacobian matrix** to exist. This is the matrix of the n^2 partial derivatives $\partial f_m/\partial y_j$, $1 \leq j, m \leq n$.
- However, it is possible that $\mathbf{y}(x)$ is *not* twice differentiable. It is possible that \mathbf{f} is continuous but does not have well-defined partial derivatives for all n^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 \mathbf{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 $\mathbf{y}(x)$ 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.